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(54) Video merchandising apparatus and method.

(57) A multimode video merchandiser system utilizes two levels of inductive learning to derive rules for selecting the sequence in which images of products stored on a videodisc are presented on a video monitor to a user. The first level of inductive learning generates rules from market survey based, consumer profile attributes assigned to items selected by previous users to determine the profile of the consumer most likely to be using the system at any given time, and to present the items in a sequence most likely to appeal to such a user. The second level of inductive learning utilizes a set of product characteristic attributes assigned to items selected by the current user to determine that user's preferences, and to modify the sequence of presentation to display first those items possessing the preferred characteristics.

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VIDEO MERCHANDISING APPARATUS AND METHOD

This invention relates to an apparatus and a method for displaying images of merchandise items to prospective customers in sequence and more particularly to such an apparatus and method which employ artificial intelligence to customize the presentation to most effectively address the prospective customer's interests.

It is becoming common today for merchandisers to utilize video systems to generate consumer interest in their products. The typical system displays images of a number of items in a fixed sequence. Some systems allow the user to select a particular group of items for review. For instance, the user may be offered the option of viewing any one of several styles of furniture. However, even in these systems the menu of selections available is fixed and the group of products selected is always presented in the same sequence. One of these known fixed menu systems will provide specifications if more detail about an item is requested, and will check a main frame computer memory to determine if a product is available, but again the order of presentation of items is fixed.

One system in use allows the user to indicate what features of a given product are most important and then selects items for presentation based on those features. For instance, if a user is interested in television sets and indicates a particular interest in 13 inch color sets, with remote control and selling for less than a given price, the system will attempt to find items meeting those specifications for display. If none are found, the least important features are successively dropped, without the knowledge of the user, until a match is found. Again, however, the sequence in which items meeting the selected specifications are presented to the consumer is always the same.

As the number of items in the system and their diversity increase, it is possible to discourage users who have particular interests, by making them view the fixed sequence of items, many of which have little appeal to them, before items are presented which may be of interest. It would be desirable therefore to be able to identify the type of person using the system in order to present first items which will most likely appeal to that type of person. However, it has been found that if persons are asked directly for information about themselves which would be helpful in determining what type of products would appeal to them, such as age, income bracket etc., the results are often not very reliable.

It has been determined that certain types of persons tend to shop at certain times. However, the trends are not so well defined that a rigid system can be reliably based upon them.

It is therefore, the primary object of the present invention to provide apparatus and a method for displaying items to the users of video merchandising systems which adapts to the pattern in which different types of persons shop.

Accordingly, with this object in view, the present invention resides in a method of controlling a video system with a programmed digital computer to display to users in a controllable sequence images of a plurality of items, said method characterized by the steps of: generating an attribute file containing as to each item an assigned value for at least one selected attribute; establishing a list of said items sorted in the sequence to be presented by the video system based upon the assigned values of said at least one attribute; controlling said video system to display said items in the order of said list; providing means for users to generate inputs indicating interest in selected displayed items; generating a data file containing as to each user input the item selected and a recurring time period during which the input was generated; determining from the attribute file the assigned value of said at least one attribute for each of the selected items in the data file; determining from the recurring time periods of the inputs and the assigned values of said at least one attribute for the selected items the most frequent value of said at least one attribute during each recurring time period; determining as to each item during the time period said input was generated the match between the assigned value of said at least one attribute and the most frequent value of said at least one attribute during that time period; and resorting said list of items in descending order of said match between the assigned value and the most frequent value during the time period said input was generated of said attribute, whereby the sequence in which said items are displayed changes in accordance with the responses of users during the respective time periods.

The invention also resides in an apparatus for sequentially displaying to a number of successive users merchandise items to which a plurality of attribute values indicative of the appeal of the item to selected classifications of users during a plurality of recurring time periods have been assigned, said system characterized by: a video recorder in which images of said items are stored; a video monitor on which said stored images are selectively displayed; interface means by which users generate inputs identifying displayed items in which they are interested; and a digital computer programmed to record the items identified by the inputs and a recurring time period during which each input was generated, to apply an

inductive learning algorithm (IL1) to the attribute values of the selected items to determine the most likely attribute values of items selected by users during each of said recurring time periods, and to select the sequence in which the images stored in the video recorder are displayed on said video monitor in descending order of a match between the attribute values assigned to each item and said most likely attribute values for the current recurring time period.

The preferred embodiment of the invention will be described by way of example, with reference to the accompanying drawings in which:

Figure 1 is a schematic diagram of a video merchandising system incorporating the invention;

Figures 2, 3, and 4 are decision trees illustrating the manner in which rules are generated in accordance with the teachings of the invention;

Figures 5a and b when placed end to end with the connections indicated by the encircled letters A and B, illustrate a flow chart of the software for the merchandiser system incorporating the invention;

Figure 6 is an illustration of a video display generated by the merchandiser system incorporating the invention;

Figure 7 is a flowchart of the software for executing the first level of inductive learning in accordance with the invention;

Figure 8 is a flowchart of the software for executing the second level of inductive learning in accordance with the teachings of the invention;

Figure 9 is a flowchart of the software for rearranging the list of items presented to the user as directed by the second level of inductive learning; and

Figures 10a and b illustrate a flowchart of the software for the rule making algorithm used by the first level of inductive learning which is described in the appendix to this application.

Figure 1 illustrates apparatus for carrying out the invention which includes a video monitor 1, a videodisc player 3, and a programmed digital computer 5. A suitable commercially available system is the Sony VIEW SYSTEM(TM) which includes their LDP 2000 laser videodisc player and SMC 2000 processor. A suitable user interface device 7 such as a touch screen attachment for the video monitor 1 provides means for the user to generate inputs to the system.

Recorded images of the items to be presented by the system, and video demonstrations of at least some of those items, may be randomly accessed by the videodisc player for display on the monitor. Sequencing of the presentations is controlled by the programmed digital computer 5 which is interactive with the user through the video monitor touch screen 7.

The system operates on the market principle that during certain hours of certain days a specific type of consumer will be shopping and that a video catalog presentation of the merchandise of interest can be customized to this type of customer. Through the use of an inductive learning program in conjunction with user input to an interactive computer, the frequency of inspection of a product is recorded when the consumer specifically requests more information about an item via an input device such as the touch screen 7. This data is then processed in an attempt to recognize viewer interest and to refine viewing patterns to present catalog items in a sequence most appealing to an individual viewer. Data from users is categorized and processed by time, day and month increments such that the video presentation will be customized to any time increment determined by user input. For example, although hourly increments may be used to categorize the inputs, through the inductive learning executions, clusters of similar hourly patterns may emerge such that three individual hours with similar presentations will seem to "merge" to form a three hour cluster. Any time increment is possible, but the exemplary system will be sensitive to hourly, daily, and seasonal increments.

The exemplary embodiment of the invention uses two separate inductive learning algorithms. The first algorithm which is used for a first level of inductive learning by the system, incorporates an algorithm called ID3 described in an article by J. R. Quinlan entitled "Induction of Decision Trees" which appears in Machine Learning 1:81-106, Kluwer Academic Publishers, Boston 1986. The second algorithm used at the second level of inductive learning is a less sophisticated program described below. Each of these algorithms requires a set of attribute values for each item of merchandise to be displayed. In the exemplary embodiment, the ID3 algorithm is used to evaluate a product gender value, a consumer income range for the product, and the age range of a consumer interested in the product. These values are the result of extensive market studies for each item of merchandise. The second less sophisticated algorithm evaluates attributes that identify the consumer's interest; such as whether the item is electronic, easy to clean, trendy, etc. It is to be understood that other or additional attributes may be utilized for the two levels of inductive learning.

The first level of inductive learning is used to determine the type of shopper who is most likely to be using the system at the present time based upon the customer attributes of the items selected by previous users during corresponding time periods. The items are then arranged by how closely their consumer attribute profiles match the assumed shopper.

- 5: In an effort to guess who the current customer is, the ID3 algorithm, as described by J. R. Quinlan, is applied to the customer attributes. Items are then arranged by how closely their consumer attribute profiles match the assumed shopper.

For the inductive learning process, several data files are used. There is a file which stores consumer profile data for each item and a separate file used to store items viewed by users. These two files are then
10 used to create data files actually accessed by the inductive learning process.

A very short example will clarify the steps involved. For the purpose of demonstration, all time parameters have been simplified. Time of day has been classified as working (w) and non-working (nw) hours, days of the week are weekdays (wd) or weekend (we) days, and months are summer (s) or non-summer (ns). These categorizations are arbitrary and have been selected for the purpose of demonstration
15 so that an "interesting" rule can be derived using a minimum of data records.

When a customer indicates an interest in an item, the system stores in a file the time of day and the product which was viewed. After several user interactions, the data file could look like this:

TABLE 1

20

User Response Data File

	Day of Week	Month	Hour	Item
25	we	s	w	Osterizer Blender
	we	s	nw	Coffee Grinder
	we	ns	w	Kitchen Center Appliance
30	wd	ns	nw	Electric Wok
	wd	s	w	Self-Sharpening Scissors
35	wd	ns	nw	Toaster
	we	ns	nw	Hand Mixer

40

Each item in the data base has a consumer attribute profile assigned through market research. The profile describes the type of person most often interested in this type of product by describing in the exemplary system such a person's age, gender, and income range. As with the time information, these attributes have been arbitrarily clustered for ease of demonstration as follows:

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TABLE 2

Consumer Attribute Categories

5	Attribute	Value Range	Value Assigned
	Age	< 25	0
10		25 - 40	1
		> 40	2
15	Gender	More Male Consumers	0
		Both Female and Male Consumers	1
20		More Female Consumers	2
	Income	< \$20K	0
25		\$20K - \$40K	1
		> \$40K	2

30

The item consumer profile for several products look like this:

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TABLE 3

Item Consumer Attribute Profiles

	Item	Age	Gender	Income
6	Osterizer Blender	(1 2 0)	(0 2 1)	(1 0 2)
10	Hand Mixer	(1 2 0)	(2 1 0)	(1 0 2)
	Coffee Grinder	(1 2 0)	(1 2 0)	(1 0 2)
15	Steam Iron	(2 1 0)	(2 1 0)	(0 1 2)
	Self-Sharpening Scissors	(2 0 1)	(1 2 0)	(0 1 2)
20	Kitchen Center Appliance	(2 1 0)	(2 1 0)	(0 2 1)
	Electric Wok	(1 0 2)	(1 0 2)	(1 2 0)
25	Toaster	(1 0 2)	(1 2 0)	(0 1 2)
	La Chef Food Processor	(1 2 0)	(1 2 0)	(1 0 2)
30	Vista Food Processor	(1 2 0)	(2 1 0)	(2 1 0)

35 The parentheses enclose an ordered list of the attributes of the consumers who most frequently purchase the product. For example, the Osterizer Blender most often is purchased by a consumer in the 25-40 year old range (designated by the 1 in the first place of the first set of parentheses), followed by the people over 40, and least by people under 25. This complete ordered listing is needed for scoring items against the assumed consumer as will be described later.

40 Using the item consumer attribute profiles of Table 2 and user response data of Table 1, three data files are built for the inductive learning process. For each record in the user response data file, an entry is created for each of the age data file, gender data file, and income data file by recording the time information and the first value in the item's consumer profile attribute value list. For example, the first user response record (we s w Osterizer Blender) would create the following three entries:

Age Data File (AGE.DAT)

45 we s w 1

Gender Data File (GEN.DAT)

50 we s w 0

Income Data File (INCOME.DAT)

55 we s w 1

The final data files for the example are as follows:

TABLE 4

Example Data File

5

AGE.DAT

10

15

20

25

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GEN.DAT

35

40

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INCOME.DAT

```

we s w 1
we s nw 1
we ns w 2
we ns nw 1
wd s w 2
wd ns nw 1
we ns nw 1

```

```

we s w 0
we s nw 1
we ns w 2
wd ns nw 1
wd s w 1
wd ns nw 1
we ns nw 2

```

```

we s w 1
we s nw 1
we ns w 0
wd ns nw 1
wd s w 0
wd ns nw 0
we ns nw 1

```

From these data files, the first level of the inductive process uses the ID3 algorithm, to derive a rule, in the form of a decision tree, that can be used to decide who is currently using the machine. According to the ID3 algorithm, the value of the information which would be gained by using a specific attribute as a branching node in the decision tree can be determined by the following formula:

$$\text{Gain}(A) = I(\text{overall}) - E(A)$$

(The gain that can be expected by using an attribute A as a branching node is equal to the information content of the attribute less the expected information requirement).

where:

$$I(\text{overall}) = -\frac{p}{p+\dots+n} \log_2 \frac{p}{p+\dots+n} - \dots - \frac{n}{p+\dots+n} \log_2 \frac{n}{p+\dots+n}$$

55

$$E(A) = \sum_{i=1}^V \frac{p_i + \dots + n_i}{p + \dots + n} I(p_i, \dots, n_i)$$

10

where A is any attribute.

p is the number of occurrences of value p for attribute A.

n is the number of occurrences of value n for attribute A.

and v is the total number of occurrences of attribute A.

15

Looking only at the gender data file, the formula would be applied as follows:

First we isolate the "classes" of the data file - or the results of the data records. In our case, the results are the consumer attribute values;

20

#occurrences of 0 = 1.

#occurrences of 1 = 4

#occurrences of 2 = 2

Calculating the information content gives the following:

25

$$I(\text{GEN}) = \frac{1}{7} \log_2 \frac{1}{7} - \frac{4}{7} \log_2 \frac{4}{7} - \frac{2}{7} \log_2 \frac{2}{7} \\ = 1.378 \text{ bits}$$

Now, isolating only the day-of-week attribute for the gender file, we get the following data table:

TABLE 5

30

Day-of-Week Attribute

Summary for Gender File

35

Day-of-Week Value

Summary of Data

we

records = 4

0 values = 1

1 values = 1

2 values = 2

40

wd

records = 3

0 values = 0

1 values = 3

2 values = 0

50

Using this data, the equation for we is:

55

$$I = \frac{1}{4} \log_2 \frac{1}{4} - \frac{1}{4} \log_2 \frac{1}{4} - \frac{2}{4} \log_2 \frac{2}{4} = 1.50$$

the equation for wd is:

$$I = \frac{0}{3} \log_2 \frac{0}{3} - \frac{3}{3} \log_2 \frac{3}{3} - \frac{0}{3} \log_2 \frac{0}{3} = 0$$

and the equation for the expected information requirement is:

$$E(\text{day-of-week}) = \frac{4}{7} I(\text{we}) + \frac{3}{7} I(\text{wd})$$

$$= .857 \text{ bits}$$

Therefore, to calculate the gain, we simply subtract the above E from the value of the overall information content calculated previously;

$$\text{Gain}(\text{day-of-week}) = 1.378 - .857 = .521 \text{ bits}$$

Similarly, the inductive learner derives:

$$\text{Gain}(\text{month}) = .414$$

$$\text{Gain}(\text{hour}) = .238$$

The ID3 algorithm chooses the day-of-week attribute as the root node for the decision making tree because it has the maximum information value. The data is then divided based upon the value of the day-of-week attribute values:

TABLE 6

Day-of-Week Decision Tree
Root Node and Resulting Data Sets

<u>Day of Week</u>						
wd			we			
ns	nw	1	s	w	0	
s	w	1	s	nw	1	
ns	nw	1	ns	w	2	
			ns	nw	2	

Notice the day of the week attribute has been removed from the data file.

The same process described above is repeated for the month and hour attributes for each of the new data record sets. After the process is complete, the rule 10 illustrated in Figure 2 results.

Similarly, rules 12 and 14 induced for the age and income data are shown in Figures 3 and 4.

Notice on the income decision tree 14 shown in Figure 4 that the final leaves of the decision trees may be ordered lists of values, rather than a single value. This occurs if "rule conflicts" arise from the data. In an ordered list, the first value is the one most often seen, the second is the next most frequent value, etc.

Once the rules have been established, they are used by the system to guess who is most probably using the system, and what kind of products that person prefers. To accomplish "best guessing", the system checks internal clocks to determine the time parameters, then executes the rules. The items in the database are then arranged according to how closely they match the rule results. For example, if a shopper enters the store at 10:00 AM on July 21, 1986, the system executes the rules for a summer weekday during working hours and determines that someone with the following attribute profile is most likely to be currently using the system:

TABLE 7

Sample Consumer Profile for Summer Weekdays During Normal Working Hours

Age - 2

Gender - 1

Income - 0

(in other words, all the retired people on Social Security are shopping now.) To re-order an item in the database, the product's consumer profile is checked against the above determined consumer profile and the items are scored by how closely the profiles match. Items with the best scores (minimum in this case) are shown first. To score the items, a simple place-checking algorithm is used. If the consumer's first profile values exactly match the item's most common consumer profile, a score of 0 results. However, if the profiles differ, 1 point is collected for each place the attributes do not match. For example, the hand mixer has the following consumer profile taken from Table 3:

	Age	Gen	Income
Hand Mixer	(1 2 0)	(2 1 0)	(1 0 2)

Checking this profile against the determined current consumer profile, we see the assumed consumer's age is one place away from the first spot, so age gets a score 1. Similarly, gender and income are both one place off and so they both get scores of 1 giving the hand mixer the composite score of 3. Similarly, all other items in the data base are scored, then the items are sorted by score in ascending order.

Utilizing this sample consumer profile and place checking algorithm, an ordered list of the sample items is generated which looks like this:

TABLE 8

Ordered List as Determined by
First Level of Inductive Learning

	Score
Self-Sharpening Scissors	0
Kitchen Center Appliance	1
Steam Iron	1
Coffee Grinder	2
Toaster	2
Le Chef Food Processor	2
Hand Mixer	3
Self Sharpening Knives	3
Osterizer Blender	4
Electric Wok	4
Gourmet Cutlery Set	4
Vista Food Processor	4

This list arranges the items in the order in which they are most likely to appeal to the assumed user of the system during the current time period selected for the example. As data accumulates, the profile of the assumed user could change based upon the selections made by previous users so that the list may be reordered based upon experience. The list will be automatically reordered to reflect both short term and long term trends in the times that different types of shoppers use the system.

The reordered list of items is used to determine the order in which items are presented to a user. The effects of reordering of the list are best seen in a browse mode of the system to be discussed below in which all of the items are displayed sequentially in the order of the list. The order, however, also influences all other item retrievals which are also discussed below. If for example, the system is directed to access
 5 drink makers, the order in which those products are shown will be consistent with their placement in the ordered list.

The reordering of the database is only carried out between users. Once a user initiates a session, the order of items in the data base remains fixed until the session is terminated. Termination of use by a given user is indicated by an input by the user so indicating, or by the elapse of a given interval following a
 70 request for an input by the system. If another user should begin using the system within the given interval so that the inputs appear to originate from the same user, there is no significant affect on the first level of inductive learning since it is the time the inputs were generated which is considered by the system and not whether they were generated by one or more users. It will, however, result in a delay in reordering of the data base by the first level of inductive learning, and will have a very definite affect on the second level of
 75 inductive learning.

The second level of inductive learning attempts to identify item characteristics which are interesting to a specific consumer. As with the first level of induction, data is collected for the personal inductive process when an item demonstration is requested. Instead of information about the item's consumer attributes, a second set of attributes which represents the user's preference in the functional characteristics of the item
 20 is tracked. If the consumer is interested in items with the same characteristics, the second level of induction will converge on that pattern and arrange the sequence of items to be shown accordingly. Data is kept only for the user session, so each user creates his or her own set of important or preferred characteristics.

In the item database, each item is related to a set of characteristics. A listing of this data using the items from the example above in order as determined by the first level of inductive learning, looks like this:

25

TABLE 9

Item Functional Characteristics

30	<u>Item</u>	<u>Characteristics</u>
	Osterizer Blender	electronic
	Hand Mixer	traditional, "easy to 35 clean", simple
	Coffee Grinder	trendy, compact
	Steam Iron	electronic, traditional
40	Self-Sharpening Scissors	traditional, simple
	Gourmet Cutlery	traditional, simple, bulky
45	Self-Sharpening Knives	trendy
	Kitchen Center Appliance	bulky, electronic, complex
50	Electric Wok	trendy, "easy to clean"
	Toaster	traditional, simple
55	Le Chef Food Processor	trendy, compact, simple
	Vista Food Processor	trendy, bulky, complex

When a demonstration of one of these items is requested, a number of things happen. First, a list of previously viewed items is checked to make sure this item has not already been processed, as items are only processed once for each consumer. If the item is a new one, the name is added to the "previously seen" list, the item's consumer attribute information is stored for the first level of induction, and then the item's characteristics are processed. For each characteristic of an item, an entry is kept in the database stating the characteristic and the number of times that characteristic has occurred this user session. For example, if the consumer views the Hand Mixer, the following entries would result:

TABLE 10

"Important Characteristics" Table
Following Demonstration of the Hand Mixer

Characteristics	# Occurrences
traditional	1
easy to clean	1
simple	1

When only one item has been demonstrated, no reordering of future presentations is done, because a characteristic is not considered significant unless it appears more than once. Thus, the above entry would cause no change in order. If, however, the consumer then looks at the toaster, the entries in the characteristic table look like this:

TABLE 11

"Important Characteristics" Table
Following Demonstration of the Toaster

Characteristics	# Occurrences
traditional	2
easy to clean	1
simple	2

Now a significant trend is developing.

An example will show how the trend affects the later presentation of items. Assume the entries in Table 11 were logged by the computer in response to the consumer's selections. Also assume the ordered list of items in Table 8 makes up the complete ordered item database. If the consumer requests to see items which chop, cut or grind vegetables, the following list of items is collected from the ordered database in Table 8 and sent to a "user-presentation" routine:

List 1

Items Which Chop/Cut/Grind Vegetables

("LeChef Food Processor" "Self-Sharpening Knives" "Osterizer Blender" "Gourmet Cutlery Set" "Vista Food Processor")

To order these items, the user presentation routine first creates the list of "important characteristics" established by looking for those characteristics which appear most frequently in Table 11. In this example, the list is:

List 2

70

"Important Characteristics"

(traditional, simple)

The item List 1 is then split into three sublists based on how many characteristics the items have in common with the list of "important" characteristics. First a sublist of items which have ALL the "important" characteristics is collected from List 1. In this case, only the Gourmet Cutlery Set is both traditional and simple. The remaining four items are then checked to see if they have ANY characteristics in common with the "important" ones. This list of partial matches contains only the Le Chef Food Processor, with "simple" as a characteristic. The remaining items are retained in their original order relative to each other based on their order of appearance in Table 8. The three sublists are then appended to give the final reordered list:

List 3

25

Ordered List Based On Second Induction

All Match	("Gourmet Cutlery Set")
+ Partial Match	("Le Chef Food Processor")
+ No Match	("Self-Sharpening Knives")
	("Osterizer Blender")
	<u>("Vista Food Processor")</u>

35

Final Ordered List

("Gourmet Cutlery Set"
"Le Chef Food Processor"
"Self-Sharpening Knives"
"Osterizer Blender"
"Vista Food Processor")

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The items are then presented to the user in this new order. The second level of Inductive learning serves to highlight specific "best guess" items by bringing them quickly to the user's view. The overall "flavor" of the item list is however maintained.

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As the user session progresses and other items are viewed, the Table 11 entries will change and a new characteristic could emerge as most important. The program is immediately sensitive to such changes, as the entire process is repeated each time a group of items is passed to the user presentation routine. The process is also rerun each time an item is demonstrated so that the remaining items will be ordered in the most up-to-date order possible. Between users, all characteristics data stored for the previous user session is removed and a new table is established for the next consumer.

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The merchandiser system in which the invention is used has been designed in layers. At the core of the system is a database of product attributes, with videodisc information, pricing, and other product specific information. Above the database is a relational network which allows the functional logic to choose and order items based on desirable attribute values. Finally the user help feature extracts information from the viewer, passes on the needed data to the rest of the system, and prepares the responses for the user.

This merchandiser system has several modes of operation which are best understood by reference to the flowchart of the system software illustrated in Figures 5a and b. Steps in the program which require user input are identified in the flowcharts by an asterisk. As indicated in block 15 of Figure 5a, when the system is idle, an "attract segment", designed to catch the eye of passers-by, is executed. The attract segment consists of a continuous loop presenting current top-selling items, sales items, and in-store specials on the video monitor. The in-store specials portion of the attract segment is a written page describing items on special which are not necessarily on the system. This allows the system to remain current, without changing the videodisc for every new product. It is during the attract segment that the first level of inductive learning, the ILI program, is executed to generate a rule for the current desirable attribute values.

When a user becomes sufficiently interested in the system to touch the screen, as represented by the block 17, he or she is taken through a brief help segment designed to familiarize new users with the symbols that are used throughout the session. Then the user is presented with a main shopping menu which offers the following options:

- 1) Help with using the system.
- 2) Browsing through the catalog.
- 3) Selecting a gift.
- 4) Finding a specific item.
- 5) Finding an item which performs a specific function.
- 6) Seeing what is on sale. And one of the following:
- 7)
 - A. Attract segment item.
 - B. "Touch the bar indicating what you would like to do".
 - C. Viewing the shopping cart items.

The user selects a segment of the program by touching the appropriate statement in the menu. The help segment 19 of the program provides information to the user on the other portions of the system. A general overview of the system and how to use the help feature is initially presented, along with a listing of the other system functions. The user then can choose which segments of the program to call up.

The browse segment 21 presents an overview of all the products on the system by sequentially presenting on the video monitor 1, at a user adjustable rate, all of the items in the system. The items are sorted prior to showing them to the user so the items believed to be of most interest are shown first. Figure 6 illustrates a typical display 23 for the browse segment. In addition to a color picture 25 stored on the videodisc of the item, the display includes symbols which enable the user to control the flow of the program. For instance if the user would like to see a demonstration of the item he touches the "demo" symbol 27. If he would like to return to the main menu he presses the "menu" symbol 29. If the user would like to save the item in his "shopping cart" he presses the "save" symbol 31 which generates the "in cart" symbol 33 indicating that the item has been saved.

The user can further control the presentation by touching the "go back" symbol 35 to return to an item that has been previously shown, the "stop" symbol 37 to freeze the currently displayed item for 20 seconds, or the "restart" symbol 39 to resume the sequencing. The items are displayed for a nominal interval, for instance 5 seconds, however, the user can speed up the sequence by pressing the "faster" symbol 41, or slow the sequence down by pressing the "slower" symbol 43. The preferred rate established by the user is stored and used as the default rate for subsequent presentations to the current user.

The gift segment 45 assists the user in selecting a gift. It is based on a marketing theory which asserts that people should be categorized by their life styles and values when choosing suitable gifts, rather than more typical things like income or age. Thus the gift suggestion segment 45 begins by showing 6 images of people and asks the user to choose which picture best describes the gift recipient. When a selection is made, a number of gift ideas, such as four, are presented which have been determined most appreciated by the group chosen. If the user is not pleased with these suggestions, the system will ask more typical questions, for example, the hobbies and age of the recipient, or a desired price range of the gift, and then additional suggestions will be made based upon these selections.

The specific items segment 47 provides the user with two means for calling up information on a specific item. The user can describe the product by typing in a product name or description which utilizes a "spell" subroutine 49, or he can choose to see an index 51. If the user types in a product name or description, a parsing routine is used to make a "best guess" 53 of what the user has requested. If the user requests a
 5 generic item, for instance blenders, all items meeting that description, and items having similar functions, will be selected for presentation.

The specific function segment 55 permits the user to request a showing of items which perform a particular function or combination of functions. This segment is also known as "build-a-sentence", because when the user chooses this option, a list of verbs and nouns are presented. When the user touches a verb
 70 describing what should be accomplished, the valid nouns for the chosen verb are highlighted. The user completes a sentence by touching the appropriate noun and all items fitting the description are used. The relational network is used to make creative as well as predictable suggestions.

The sales segment of the program 57 allows the viewer to browse an electronic sales flyer. Any items on the system which have been marked "on sale" by store personnel are presented, in an order determined
 15 by the inductive learner. When all the items on the videodisc have been shown, the user has an option of seeing "other in-store specials". These are short, appealing descriptions of other products in the store which should be noticed, but which are not yet on the system.

The last option, 7, presented to view on the main menu is dependent on the user's path through the system, and whether any items have been "saved" until now. When the user is first shown the main menu,
 20 after touching the screen during the attract segment, the final option given the user is to see the item that was on the screen during the attract segment when the user touched the screen. For example, if the user saw the Gourmet Cutlery Set and decided to initiate a user session by touching the screen, the final option would be:

"or touch here to see the Gourmet Cutlery Set."

25 This option is offered so that the user does not have to relocate the item which interested him in initiating a session. The user only has one chance to respond to this option, however, as one of the following options is presented in subsequent presentations of the main menu.

If the user is not seeing the menu for the first time, the final option may in fact be just a statement directing user to "touch the bar indicating what you would like to do". This statement merely acts as an
 30 instruction, reminding the person to touch the screen to provide input.

The last possibility for the final option on the main menu is to "see the items in your shopping cart." When the user chooses to view the items in the shopping cart, the shopping cart segment 59 of the program is entered. In this segment, a screen displaying all the items which have been saved is displayed. The user can then see a demonstration of the item, or choose items to be included in a comparison as
 35 indicated in block 61. All the items specified will be compared and described 63. The first comparison prepared is a pricing comparison. In addition to the usual pricing information, the pricing comparison tells if there are multiple models of an item, and whether an item is on sale. Following the price information, the functional comparison of the items is prepared. For this comparison, items with similar functions are grouped together and a summary of the results is printed on the screen. Finally, the user can see a detailed
 40 description/comparison of the items. Again, like items are grouped, but this time the features are highlighted, first as to how the products are similar and then as to how they differ. If the items chosen from the shopping cart are not comparable, individual descriptions are supplied. All comparisons are created during the user's session, not preprogrammed, so that any changes in product features or pricing information would be included in reports without any programming changes required.

45 As mentioned previously, the merchandiser system has been designed in layers. The only layer which the user sees is the top-most user help feature or the "communicator". The communicator performs many tasks, all designed to make the system more user-sensitive, responsive, effective, and generally more "intelligent".

The most visible task which the communicator performs is the selection and offering of help to the user.
 50 If the user takes longer than is "normal" to respond to a menu or choice, the communicator will select a helpful message based upon the user's place in the system and the path taken to reach that place, and print out the message in a "FYI", (For Your Information) window. The system will make three such suggestions, telling what the various options will do, or even suggesting an appropriate response. If no response is noted after the third message, the system assumes the user has left and "times-out", re-
 55 executing the first level of inductive learning, IL1, and beginning the attract segment 15.

In addition to offering helpful suggestions to the user, the communicator performs background functions which influence the user, but which are not obvious. These functions are performed by the portion of the software described by the flowchart of Figure 5b. After the user branches to any of the five main user options (browse, gift, specific time, specific function, or sale items) the communicator accesses the relational network to choose appropriate items, block 67, and then rearranges those based upon the second level of inductive learning, IL2, to make a "best guess" as to what will be most interesting to this consumer 69. Rearrangement of the items at this point in the program is only carried out if the current user has made at least two selections. For instance, if the user has just entered the system and has requested the browse segment, there is no information available yet to determine this user's preferences, and hence this rearranging subroutine is bypassed.

Appropriate items for the mode selected by the user are displayed one at a time as indicated in block 71 in the sequence determined by the first level of inductive learning, and the second level of inductive learning, if appropriate. A request by the user, block 73, for a demonstration of the item currently being displayed on the video monitor is used as an indication of the user's particular interest in a specific item. Thus, in response to a request for a demonstration, the second level of inductive learning IL2 program is run, block 75, while a video recording stored on the videodisc of a demonstration of the product is presented on the video monitor, as indicated in block 77. The IL2 program, which is discussed more fully in connection with Figure 8, records the fact that the demonstration has been requested on the item, together with the date, time, and month of the entry. Upon completion of the IL2 program, the remaining items selected for the specified system mode are rearranged in block 81 based on this updating of the second level of inductive learning, if appropriate.

Whether or not a demonstration has been requested, the shopper has the option of saving the item in his or her electronic shopping cart, block 83. Items saved by touching the save symbol 31 are stored, block 85, for further consideration.

At any time, the user has the option, block 87, of pressing the menu symbol 29 if he or she has seen enough items in the present mode and desires to return to the main menu. In the absence of an indication that the user wants to return to the main menu, the program loops back, block 89, to present the next item in the selected list of items.

When all of the selected items have been presented, a submenu is generated, block 91, presenting the options available to the user. The options available depend upon the user's past activity. For example, a user completing the gift section will be offered the option to review the gift suggestions presented, or see different gift suggestions. If the user has saved items in the shopping cart, the option will be given to review these items also. To enhance the versatility of this menu, the program is written to interpret the sentence which describes the option chosen by the user. This is done using a parsing routine, which, in the exemplary embodiment, is a key word matching routine which analyzes the character strings and causes the program to branch to the appropriate routine. This allows the options to be reordered, reworded, or new options to be added without changing any code.

Figure 7 illustrates the flowchart for the first level of inductive learning, IL1. When the system times out, indicating that the last user has terminated use of the system, the ID3 routine for the selected attributes, income, age, and gender in the exemplary embodiment, is run, block 93. A detailed flowchart for the ID3 routine is illustrated in commonly owned U.S. patent application Serial No. 826,524 filed on February 8, 1986 in the names of William J. Leech and Thomas N. Dean and entitled "An Automated Rule Based Process Control Method with Feedback and Apparatus Therefore". For convenience, this flowchart is repeated and discussed in the appendix of this application. Software implementing the ID3 algorithm is commercially available from the assignees of this invention, Westinghouse Electric Corporation, under the trademark MasterMinder. Herman Edge of Palo Alto, California, also markets software implementing the ID3 algorithm under the designation Expert Ease.

The rules derived by the ID3 routine for income, age and gender are loaded in block 95. Each of the rules is executed in block 97 to determine a current consumer profile by applying the rule to the data stored in Table 1 in the manner described previously. Each of the items is then scored in block 99 according to how closely they match the profile and the items are then sorted based on the score in block 101 and the sorted list is stored in block 103. If a new user has not touched the screen, block 105, the attract segment items are presented in block 107. When a new user does enter the system, the introductory page is presented, block 109, to instruct the user on how to use the system.

Figure 8 illustrates in more detail the flowchart of the IL2 routine illustrated by the block 75 in Figure 5b. First a check is made in block 111 to determine if the item has already been processed to prevent counting it twice. If the item has already been recorded for this user, the remainder of the IL2 routine is bypassed and the demonstration is played in block 77. New items for this user session are added to the item list in

block 113 and the consumer attribute information for the item is recorded for the first level of inductive learning, IL1, in block 115. If the item being examined does not have any assigned functional characteristics as determined in block 117, the video demonstration is played. If the assigned functional characteristic is already listed in the table as determined in block 119, the number of occurrences of that characteristic is incremented in block 121. If it is a new characteristic, it is added to the table with the number of occurrences set equal to 1, as indicated in block 123. Since an item may be assigned more than one characteristic, this procedure is repeated for each characteristic assigned to the item under review. When all of the characteristics have been recorded, or if no functional characteristic has been assigned to the item, the IL2 run routine is completed and the video demonstration is played in block 77.

Figure 9 illustrates a flowchart for rearranging items in accordance with the second level of inductive learning, IL2. First the list of items to be presented is gathered in block 125. Next, a check is made in block 127 to determine if there are any trend candidates, that is, are there any entries in the occurrence of characteristic tables. If not, the list of items is presented as a final list in block 129 with the sequence unchanged. If there are any entries in the characteristic list, they are sorted by the number of occurrences in block 131. If none of the characteristics has occurred more than once as indicated in block 133, no significant trend has been detected and the list is not rearranged. If any characteristic has occurred more than twice, a list is made in block 135 of the top candidates. That is, a list of the characteristic or characteristics that have occurred the most number of times. Such characteristics are considered the important characteristic or characteristics. Next a list is made, block 137, of the items, if any, which match all of the important characteristics. A list is then made in block 139 of all of the items which have a partial match with the important characteristics. That is those items which are assigned at least one of the important characteristics. Finally a list is made, block 141, in which the item having a total match with the important characteristics followed by those items having a partial match are moved to the top of the list of selected items as ordered by the first level of inductive learning, IL1. Items for which there is no match with the important characteristics remain in the sequence determined by IL1. This rearranged list is then presented as the final list in block 129.

A number of the features discussed above make this system unique. First is the use of inductive learning techniques to teach the system items of interest by determining the most desirable attributes at any given time based upon the input of previous users. This allows the system to be truly sensitive to user interest and to respond appropriately. One significant advantage of this is that even after a pattern has developed, e.g. a trend in shopping habits changes for some reason, the inductive learning program will detect this change and adjust the presentation accordingly. In addition, the system can tailor a presentation to the individual taste of the current user, again using an inductive learning program which is based upon a set of assigned attributes.

In addition to being responsive, the system is highly versatile. The versatility is achieved by reducing the product data base to a series of attributes based on extensive market research. The products are retrieved solely by their attributes, permitting items to be added or deleted, or the whole data base to be changed, without significantly changing the computer code. This allows the system to be generic in that any type of items may be added to the catalog, and they will be retrieved or sequenced using only their attribute values. Versatility is further enhanced by the system by the extensive use of automatic screen generation techniques such that most screens shown to the user are generated based upon the current user situation, making wording changes, option changes and formatting changes trivial to implement.

APPENDIX

A detailed flow chart of the ID3 program suitable for generating the decision trees which derive the rules for the first level of inductive learning used in the present invention is set forth and described in commonly owned United States Patent Application Serial No. 826,524 filed on February 8, 1988 in the names of William J. Leech and Thomas N. Dean and entitled "An Automated Rule Based Process Control Method With Feedback And Apparatus Therefore". This flow chart, which appears as Figures 8a and 8, and the description thereof which appears on pages 18-20 of that application, are repeated here as Figures 10a and b and the description below for the purpose of providing suitable means for implementing the ID3 rule making algorithm.

Once the sample set is selected, new rules are induced as illustrated in Figure 10. First, the samples are input 112 and stored in a data structure, such as an array, matrix or list which is suitable for storing samples having multiple attributes where the samples must be sorted and subdivided during processing. Next, an attribute is chosen 114 for testing to determine whether it is an appropriate split node for the

decision tree. The attribute to be tested can be chosen arbitrarily, such as choosing the first attribute in the sample. The attribute is tested 116 to determine whether it is numeric. Assuming that the attribute is numeric, a determination 118 is made concerning whether the attribute has a value in each sample. If each attribute does not have a value, a suitable value for that sample is generated 120 by, for example, averaging the attribute values or randomly picking an attribute value. Another strategy for handling absent values is to create pseudosamples to substitute for the sample with the missing attribute. The missing attribute value for the pseudosamples can be generated based on the distribution of the attribute values for other samples in the sample set or on a distribution such as a bell curve.

The records are sorted 122 in ascending sequence according to the attribute value for the attribute selected, and then the sample set is split 124 into two sets between the first and second samples producing a split value and a split index. Next, an information gain function for creating a decision tree is evaluated 126 to produce the information gain for the split previously selected. The information gain will be used to determine whether the tested attribute is appropriate as the split node for the current level in the decision tree. The calculations necessary to evaluate this preferred function will be discussed in detail later. The resulting information gain along with split values and split index are stored 128. If the last sample has not been reached 130, the index is advanced 132 and the sample set is split between the next two samples and the decision tree generation function is again evaluated. Once the end of the samples is reached, the split which produces the maximum information gain, along with the split value and split index, is saved 134. If the end of the attributes has not been reached 136, then the next attribute is chosen and the process continues again.

After a similar process is performed for each symbolic attribute, the numeric or the symbolic attribute having the highest information gain and the corresponding indices are chosen 158 by comparing the information gain for each attribute. Next, the maximum numeric and symbolic information gain values are compared 160 with each other, the attribute with the maximum is chosen and a decision tree record entry is created. The data structure of the decision tree will be discussed in more detail hereinafter.

After the maximum is chosen, the samples are again sorted 162, based on the selected attribute, into ascending sequence if numeric and according to symbolic name if symbolic. Then the sample set is divided into subsets based on the attribute with the maximum information gain and the corresponding split value and split index. If all the samples in each subset have the same class values, then the decision tree for the subset structure and the corresponding rule is saved. If each sample in a subset does not have the same class values, then the algorithm is then executed for the subset as if it were the set. That is, the subset is treated as a set and the process of choosing attributes, sorting into ascending sequence, splitting and producing the information gain is executed.

If during the execution of the above-discussed rules induction algorithm an attribute is encountered which requires the solution of a subproblem to produce a subrule, the subrule is produced in the same manner as previously described. In the preferred embodiment, the subproblems which appear in the primary problem are rule induced prior to primary problem rule induction, so that when a subproblem indicator is encountered, the rule for that subproblem can be made available immediately. An alternative to performing subproblem induction before primary problem induction, is to temporarily halt rules production based on the primary sample set while a rule is produced for a subproblem sample set. Either the preferred or alternate approach can be applied whenever a particular class value requires the solution of a subproblem.

45 Claims

1. A method of controlling a video system (1, 3) with a programmed digital computer (5) to display to users in a controllable sequence images of a plurality of items, said method characterized by the steps of:
 - generating an attribute file containing as to each item an assigned value for at least one selected attribute;
 - establishing a list of said items sorted in the sequence to be presented by the video system (1, 5) based upon the assigned values of said at least one attribute;
 - controlling said video system (1, 3) to display said items in the order of said list;
 - providing means (7) for users to generate inputs indicating interest in selected displayed items;
 - generating a data file containing as to each user input the item selected and a recurring time period during which the input was generated;
 - determining from the attribute file the assigned value of said at least one attribute for each of the selected items in the data file;

determining from the recurring time periods of the inputs and the assigned values of said at least one attribute for the selected items the most frequent value of said at least one attribute during each recurring time period;

6 determining as to each item during the time period said input was generated the match between the assigned value of said at least one attribute and the most frequent value of said at least one attribute during that time period; and

resorting said list of items in descending order of said match between the assigned value and the most frequent value during the time period said input was generated of said attribute, whereby the sequence in which said items are displayed changes in accordance with the responses of users during the respective
70 time periods.

2. The method of Claim 1 further characterized in that the step of determining the most frequent value of said at least one attribute comprises generating a rule (10, 12, 14) from the recurring time periods in which the inputs are generated and the values of said attributes assigned to the selected items, and which rule groups together recurring time periods in which the most frequent values of said at least one attribute
75 are the same.

3. The method of Claim 2 further characterized in that the step of generating said attribute file comprises generating such a file containing as to each item an ordered group of assigned values for said at least one attribute, wherein said step of determining the most frequent value of said at least one attribute during each recurring time period utilizes the highest order value of said attribute for the selected items,
20 and wherein said matching step takes into account the order in which the most frequent value is assigned to said attribute for each item.

4. The method of Claim 3 further characterized in that each ordered value of each attribute for each item is assigned a numerical score, wherein the matching step comprises determining as to each item the numerical score assigned to the most frequent value of the attribute during the current time period, and
25 wherein the resorting step comprises arranging the items in order of the numerical score of the most frequent value of the attribute for the current recurring time period.

5. An apparatus for sequentially displaying to a number of successive users merchandise items (25) to which a plurality of attribute values indicative of the appeal of the item to selected classifications of users during a plurality of recurring time periods have been assigned, said system characterized by:

30 a video recorder (3) in which images of said items are stored;

a video monitor (1) on which said stored images are selectively displayed;

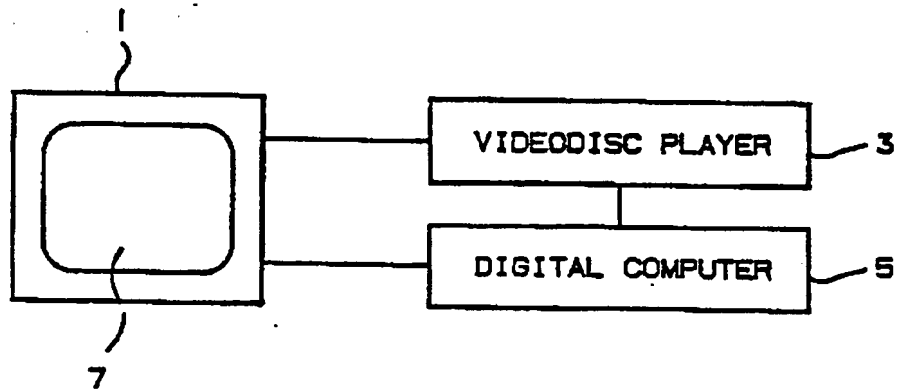
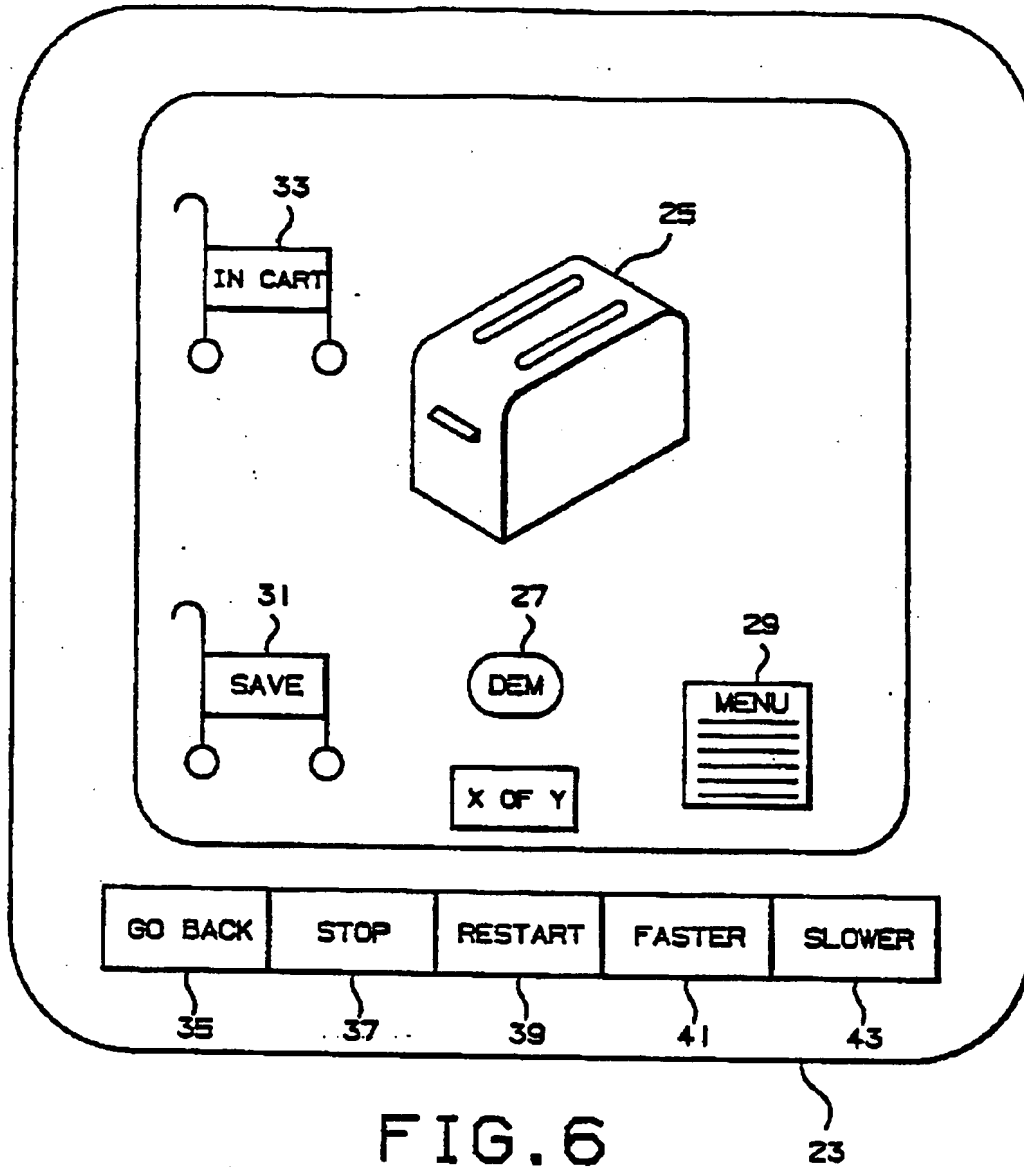
interface means (7) by which users generate inputs identifying displayed items in which they are interested; and

a digital computer (5) programmed to record the items identified by the inputs and a recurring time
35 period during which each input was generated, to apply an inductive learning algorithm (IL1) to the attribute values of the selected items to determine the most likely attribute values of items selected by users during each of said recurring time periods, and to select the sequence in which the images stored in the video recorder (3) are displayed on said video monitor (1) in descending order of a match between the attribute values assigned to each item (25) and said most likely attribute values for the current recurring time period.

40 6. The system of Claim 5 further characterized in that said merchandise items (25) have assigned item characteristics and said computer (5) is programmed to apply a first inductive learning algorithm (IL1) to said attribute values of items identified by inputs by all users, to apply a second inductive learning algorithm (IL2) to said item characteristics assigned to items selected by the present user to determine the item characteristics most appealing to the present user, to adjust said selected sequence in which the items are
45 displayed to the present user only, and to present first to the present user items assigned the item characteristics most appealing to the present user.

50

55



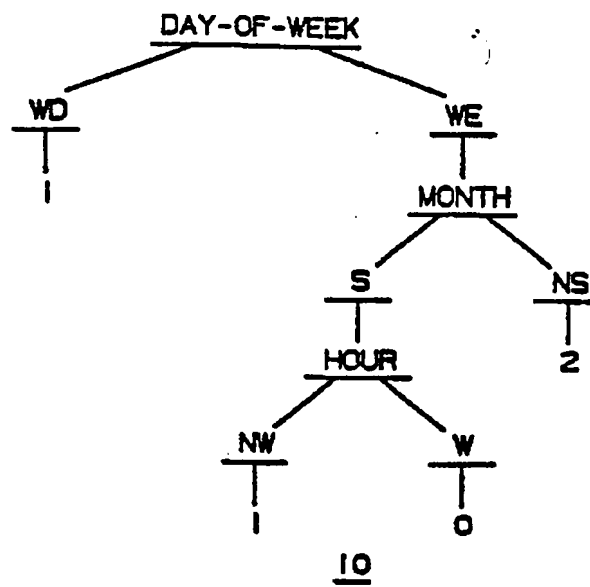


FIG. 2

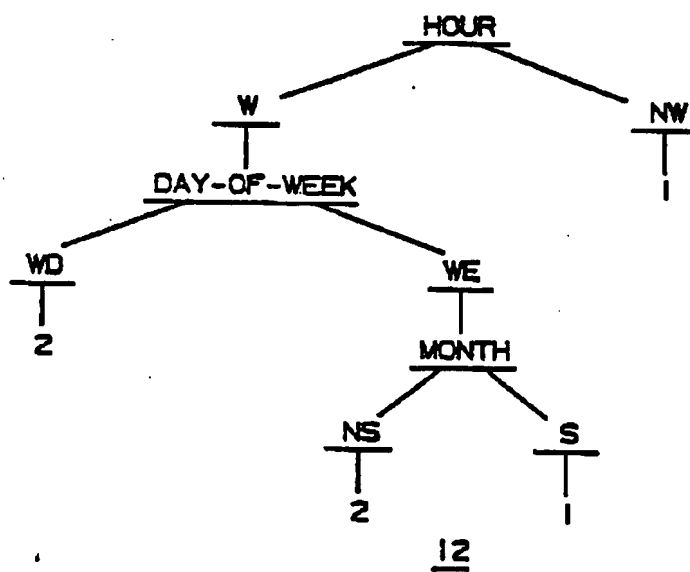


FIG. 3

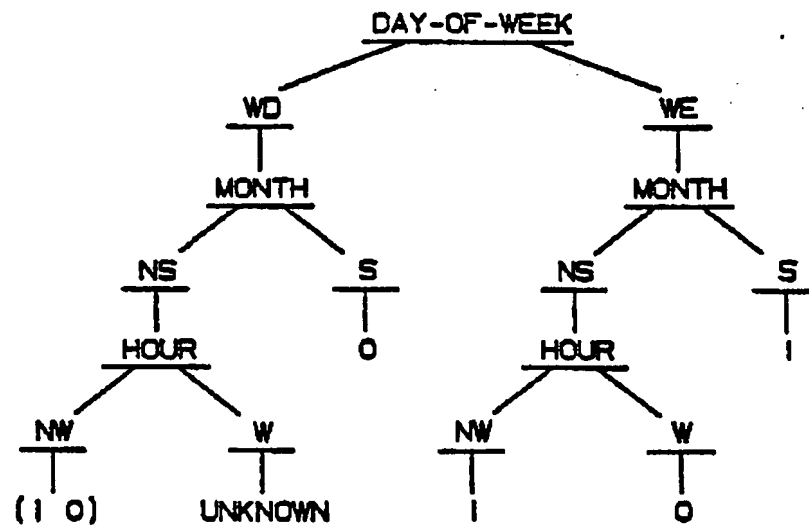
14

FIG. 4

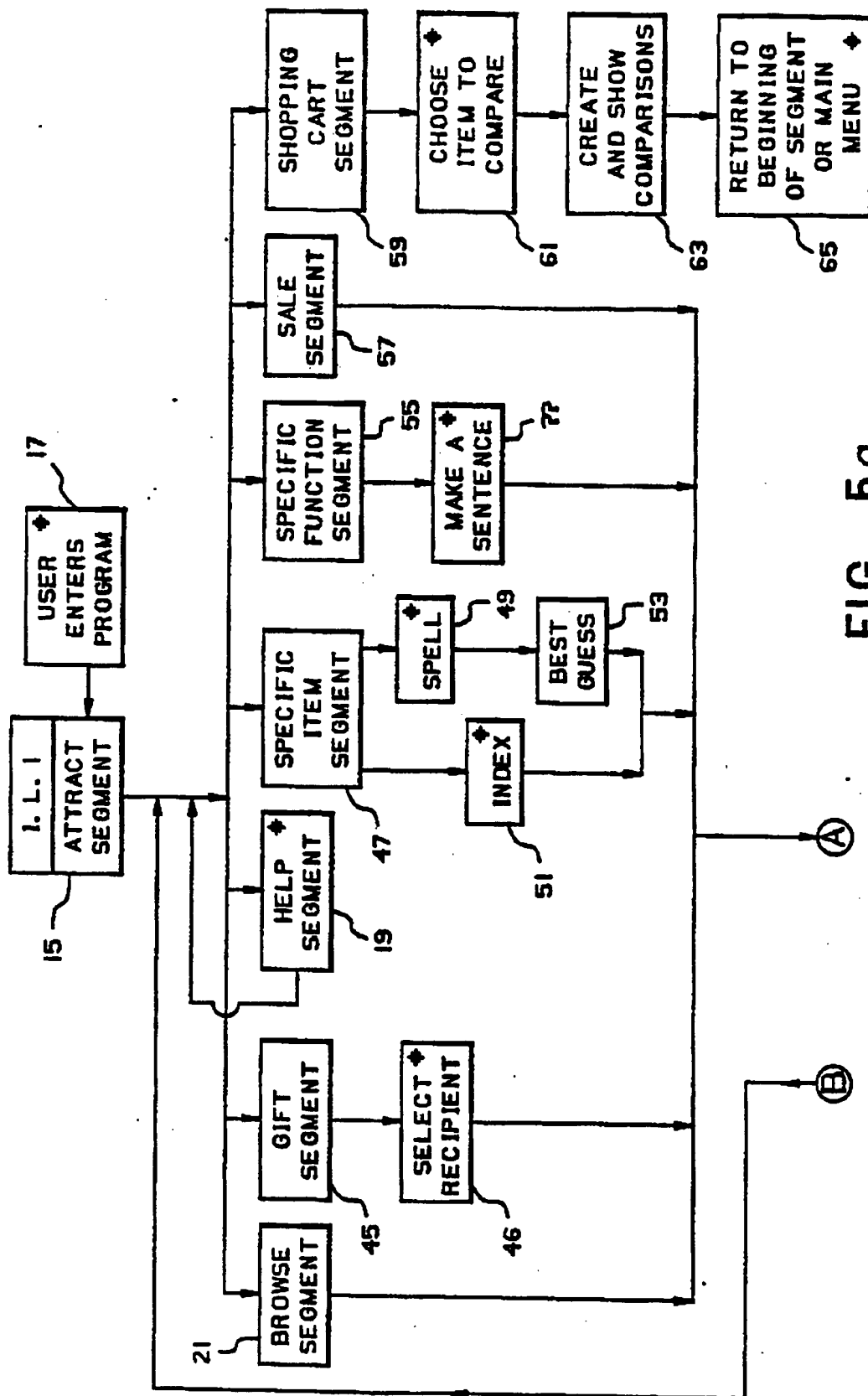


FIG. 50.

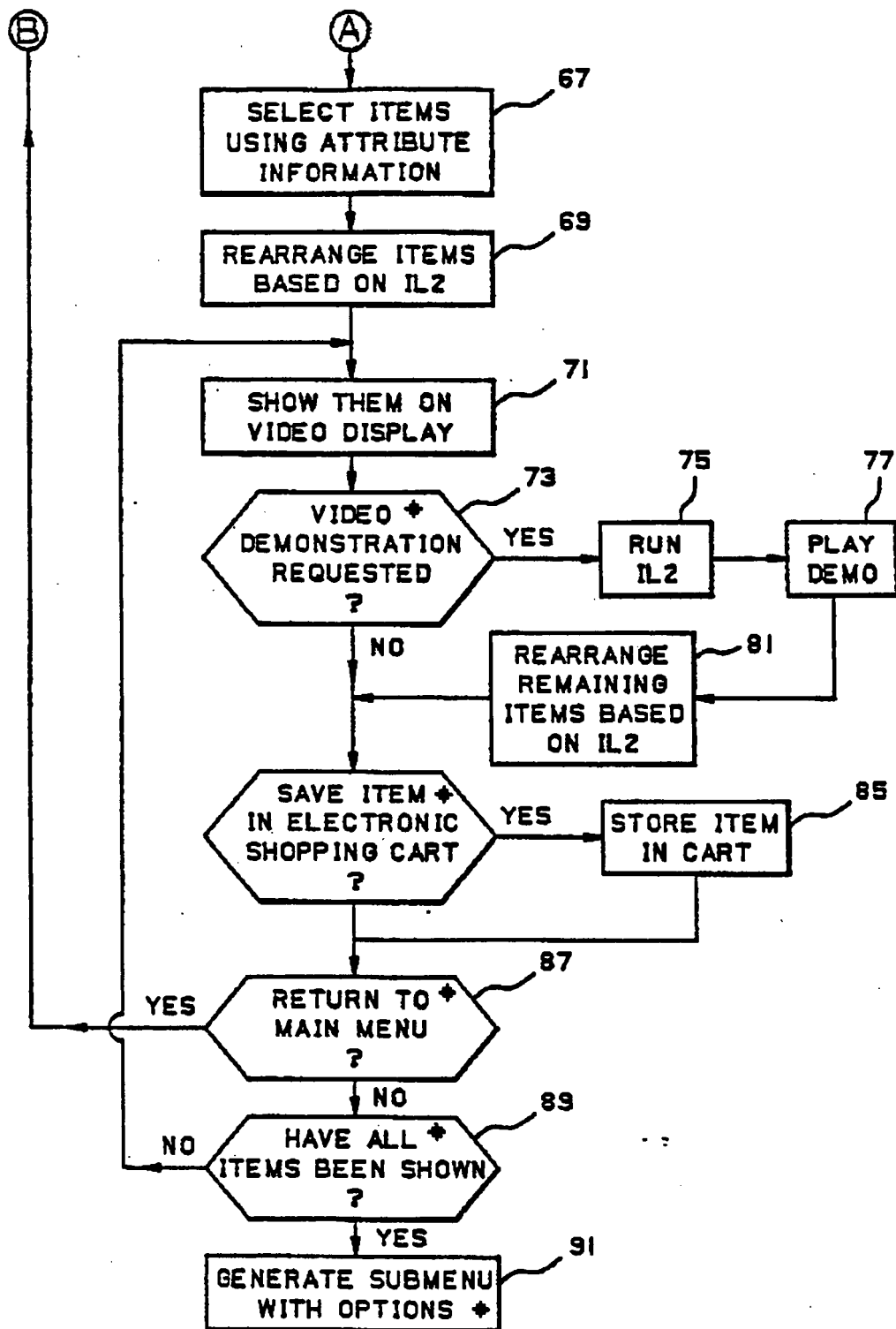


FIG. 5b.

(PREVIOUS USER TIMES OUT)

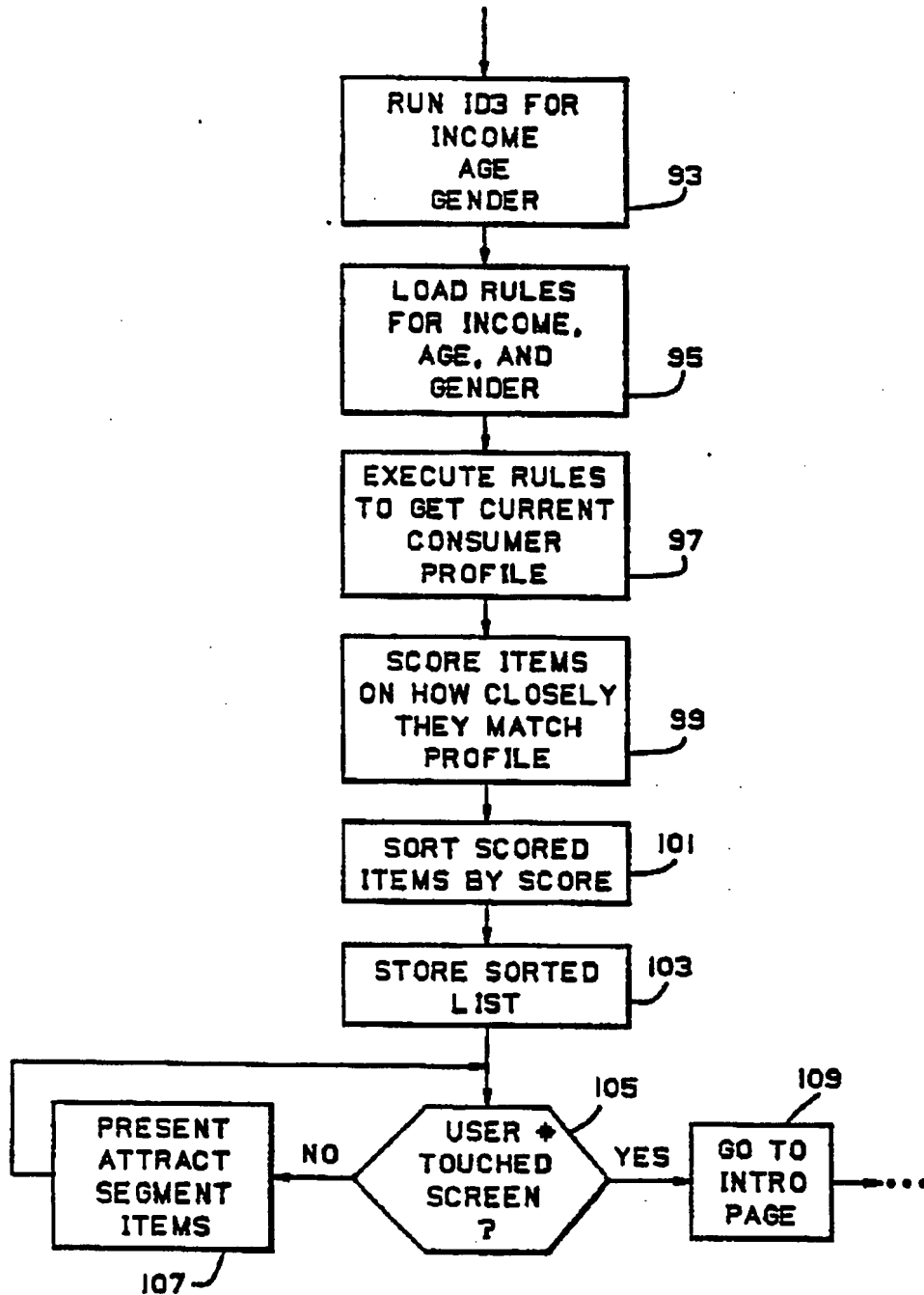


FIG. 7.

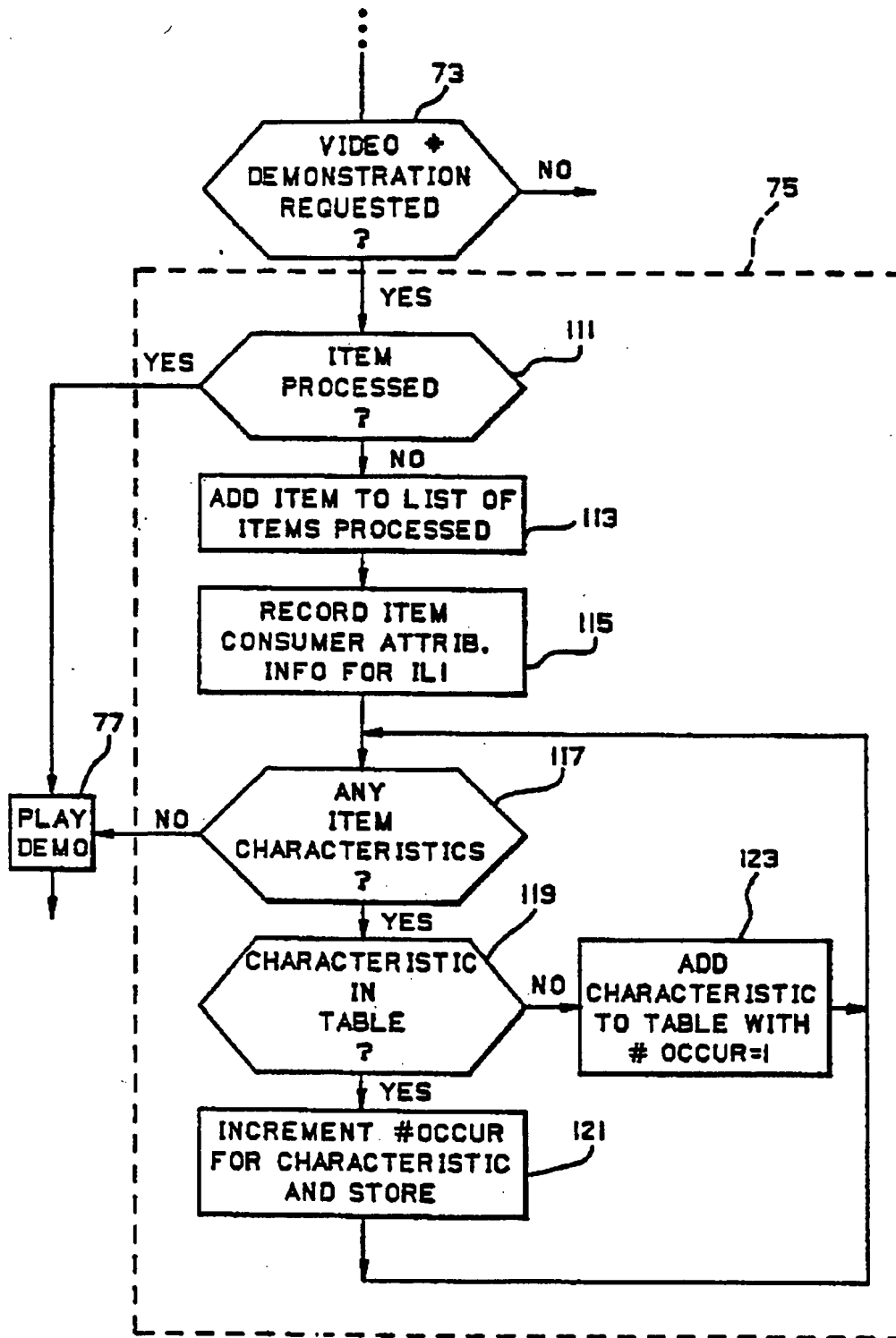


FIG. 8.

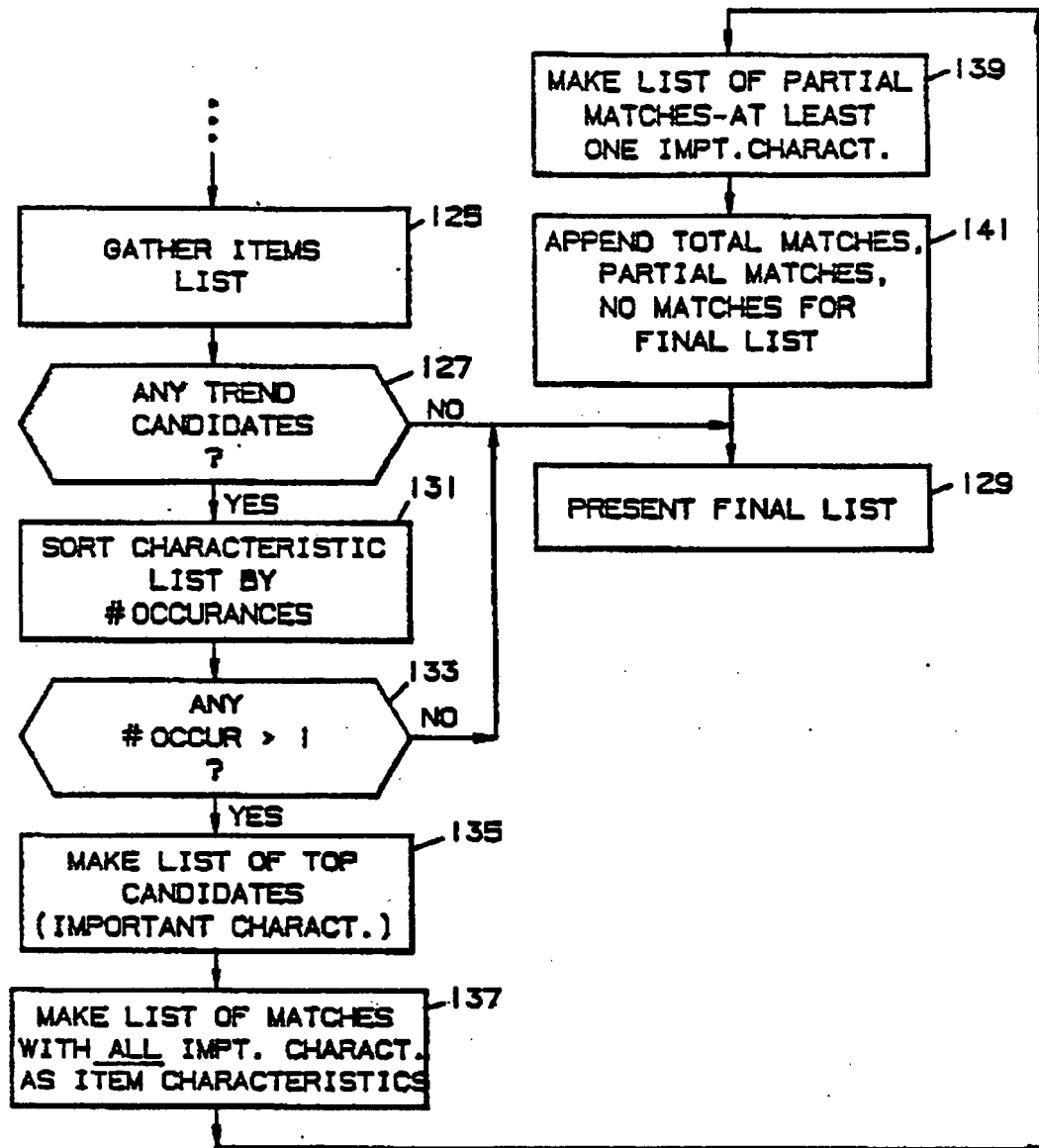
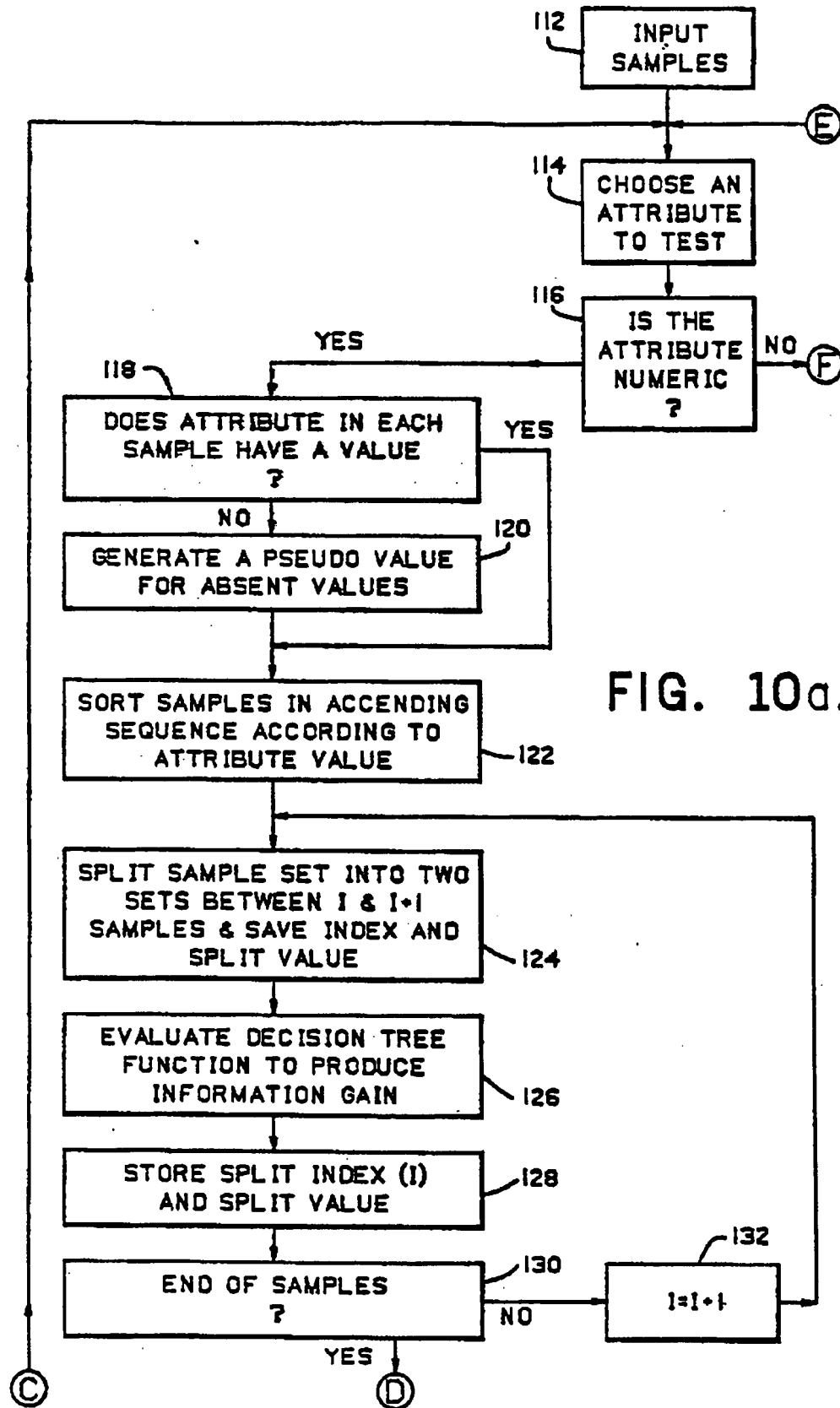


FIG. 9.



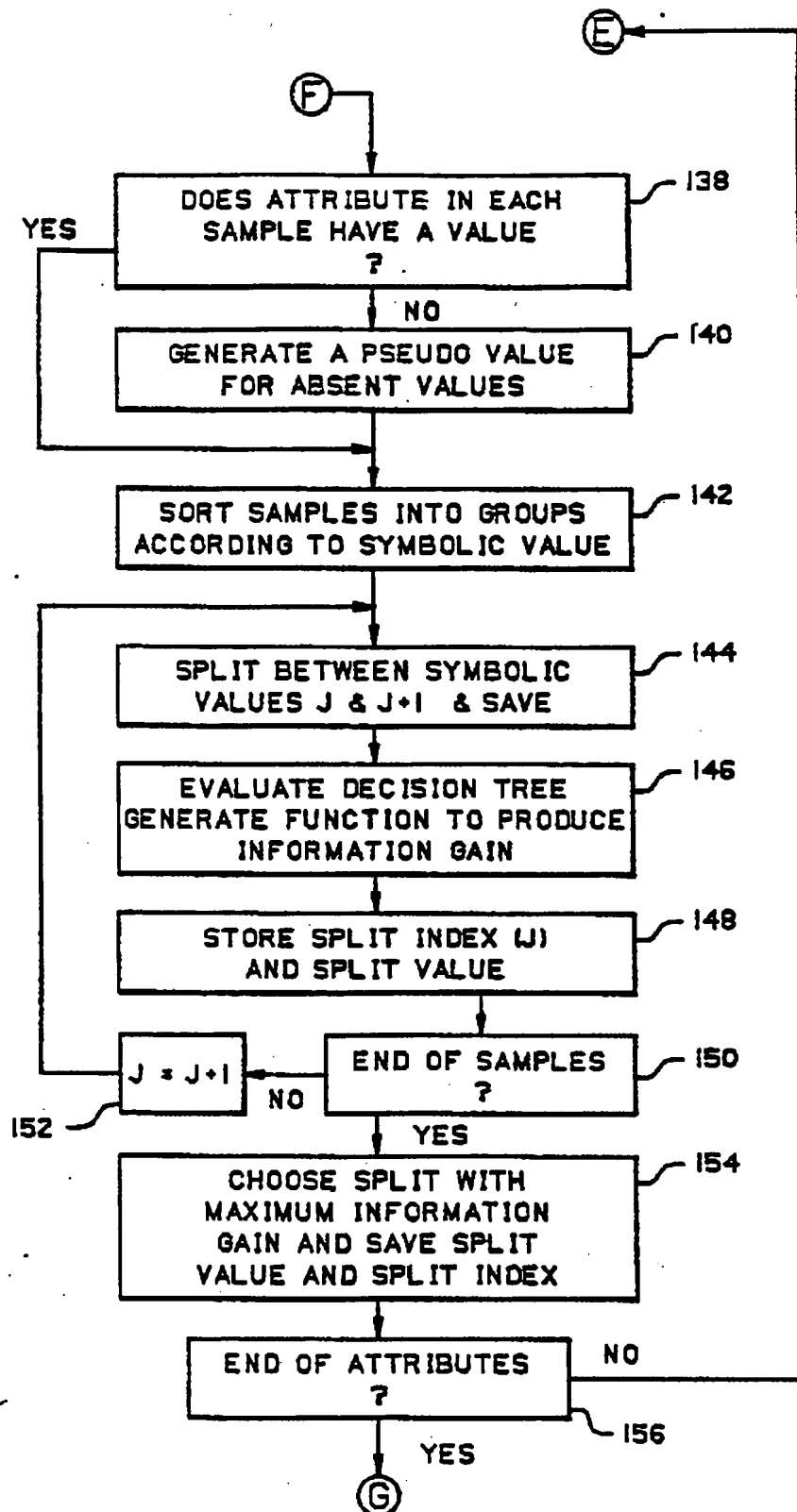


FIG. 10b.

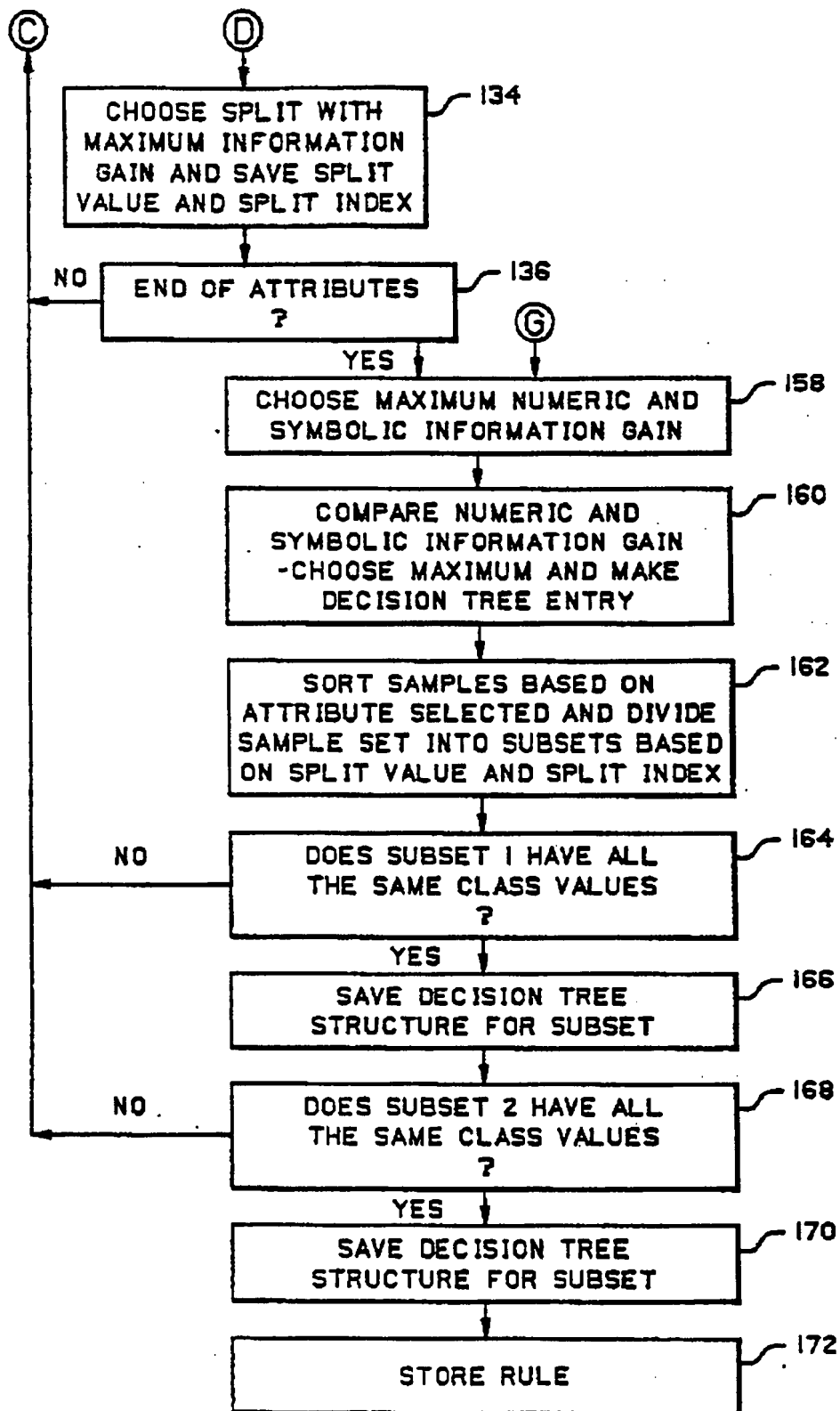


FIG. 10c.



European Patent
Office

EUROPEAN SEARCH REPORT

Application Number

EP 87 30 8403

DOCUMENTS CONSIDERED TO BE RELEVANT			
Category	Citation of document with indication, where appropriate, of relevant passages	Relevant to claim	CLASSIFICATION OF THE APPLICATION (Int. Cl.4)
A	JOURNAL OF THE AMERICAN SOCIETY FOR INFORMATION SCIENCE, vol. 37, no. 3, May 1986, pages 123-135, New York, US; MUKHOPADHYAY et al.: "An intelligent system for document retrieval in distributed office environments" * whole document *	1	G 09 F 27/00 G 06 F 15/21
A	PATENT ABSTRACTS OF JAPAN, vol. 9, no. 262 (P-398)[1985], 19th October 1985; & JP - A - 60 110 078 (FUJI) 15-06-1985	5	
A	FR-A-2 524 180 (LEQUAD) * page 1, line 19 - page 2, line 23; page 2, line 36 - page 3, line 14; page 5, line 8 - page 6, line 5; claims 1-5, 8-10; figure 1 *	5	
			TECHNICAL FIELDS SEARCHED (Int. Cl.4)
			G 09 F G 09 B G 06 F
The present search report has been drawn up for all claims			
Place of search BERLIN		Date of completion of the search 16-12-1987	Examiner BEITNER M.J.J.B.
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